Ecologic Determinants of West Nile Virus Seroprevalence among Equids, Brazil

Edmilson F. de Oliveira-Filho,¹ Carlo Fischer,¹ Beatrice Sarah Berneck, Ianei O. Carneiro, Arne Kühne, Angelica C. de Almeida Campos, Jorge R.L. Ribas, Eduardo Martins Netto, Carlos Roberto Franke, Sebastian Ulbert, Jan Felix Drexler

Among 713 equids sampled in northeastern Brazil during 2013–2018, West Nile virus seroprevalence was 4.5% (95% CI 3.1%–6.3%). Mathematical modeling substantiated higher seroprevalence adjacent to an avian migratory route and in areas characterized by forest loss, implying increased risk for zoonotic infections in disturbed areas.

West Nile virus (WNV) is a widely distributed arthropodborne flavivirus transmitted predominantly by *Culex* mosquitoes (1). Among infected persons, ≈20% show clinical signs, such as mild fever, rash, joint pain, headache, vomiting, and diarrhea (1,2); ≈0.7% have severe illness, such as encephalitis, meningitis, acute flaccid paralysis, respiratory failure, and even death (1). Beyond vectorborne transmission, transfusion-transmitted WNV infections have endangered blood safety (3). Equids are susceptible to WNV and develop severe disease (fatality rate ≤30%), are exposed to WNV vectors outside and in stables, and are spatially distributed near human settlements. Thus, equids can be sentinels for early detection of regional WNV activity (4).

In the Americas, WNV gained attention after its rapid spread in the United States beginning in 1999 (4). In South America, WNV dispersion is poorly understood. Seropositive horses were found in Colom-

Author affiliations: Charité-Universitätsmedizin Berlin, Berlin, Germany (E.F. de Oliveira-Filho, C. Fischer, A. Kühne, A.C. de Almeida Campos, J.F. Drexler); Fraunhofer Institute for Cell Therapy and Immunology, Leipzig, Germany (B.S. Berneck, S. Ulbert); Federal University of Bahia, Salvador, Brazil (I.O. Carneiro, E.M. Netto, C.R. Franke); Bahia State Agricultural Defense Agency, Salvador (J.R.L. Ribas); Sechenov University, Moscow, Russia (J.F. Drexler); German Centre for Infection Research, Berlin (J.F. Drexler)

DOI: https://doi.org/10.3201/eid2709.204706

bia in 2004 (5) and in Argentina in 2006 (6). In Brazil, the largest country in South America, serologic studies from central, southeastern, and northeastern regions suggested WNV circulation among horses since at least 2009 (7,8). Human WNV infection was described only once, in 2014, from a patient in northeastern Brazil with encephalitis (9). In 2018, a WNV strain was isolated and sequenced during an epizootic in horses in the southeastern coast (10). The horse-derived virus from Brazil clustered with strains detected in different birds in the United States in 2002 and 2005 (10), indicating that migratory birds could play a role in WNV transmission in Brazil.

Serologic WNV data from equids along avian migratory routes are scarce. In the only available study from northeastern Brazil, 1/88 horses was WNV seropositive with a low neutralization titer (7). In the absence of testing for cocirculating flaviviruses, a low WNV antibody titer could be caused by infections with other flaviviruses, eliciting cross-reactive antibodies (11). We conducted a seroepidemiologic study among equids to investigate the spread of WNV in northeastern Brazil.

The Study

We collected serum samples from 713 equids, including horses and mules, sampled as part of routine veterinary surveillance activities during 2013–2018 in the state of Bahia in northeastern Brazil. The animal ethics committee of the Federal University of Bahia approved the sampling and analyses (authorization no. 55/2017). Sampling covered a large area that connects the location of the human case from 2014 and the 2018 horse epizootic (9,10). The area is adjacent to the Atlantic, northeastern, and central avian migratory routes (Figure 1).

¹These first authors contributed equally to this article.

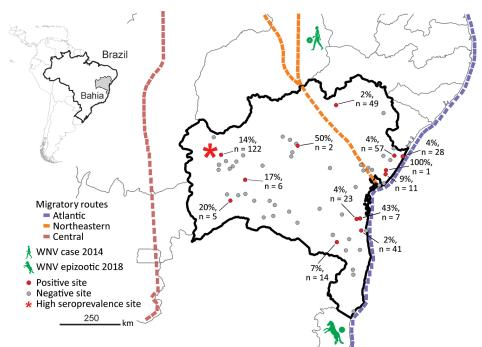
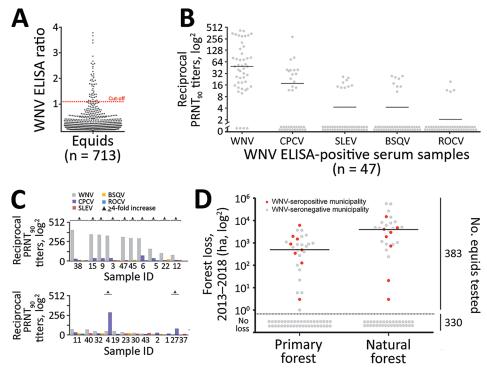


Figure 1. Geographic distribution and PRNT₉₀-validated West Nile virus seroprevalence among equids per sampling site in Bahia State, Brazil. Inset shows location of Bahia State in northeastern Brazil. Sample sizes are shown only for locations with seropositive animals. Avian migratory routes are based on the 2016 annual report of the Chico Mendes Institute for Biodiversity and Conservation (https://www.researchgate. net/publication/292980285 Annual Report of Flyways and Priority Areas for Migratory Birds in Brazil Relatorio anual de rotas e areas de concentracao de aves migratorias no Brasil). PRNT₉₀, 90% plaque-reduction neutralization tests.

Figure 2. WNV seroprevalence among equids, Brazil. A) ELISA absorbance values displayed as sample to cutoff ratio, as previously described (2). We increased the ELISA cutoff by 10% above which samples were considered positive to maximize specificity because the ELISA was not originally validated for horses in Latin America, which are infected by more Japanese encephalitis serocomplex viruses compared with horses in Europe. Dotted orange line represents the 1.1 positivity cutoff. B) Reciprocal PRNT on titers for WNV and other flaviviruses. Statistical significance levels were inferred by using the Kruskal-Wallis test. Bars indicate mean. Graph created by using Prism (GraphPad software, https:// www.graphpad.com). C) Distinction of heterotypic serum samples based on the endpoint titers of various flaviviruses. Triangles indicated endpoint



titers ≥4-fold. D) Effects of forests and forest loss on WNV seropositivity and seronegativity among equids in municipalities, Brazil. Natural forest is made up of introduced or native tree or vegetation that have reproduced naturally, without help or (human) intervention. Primary forest is made up of intact and nonintact natural forest and refers to areas that reached the final stage of succession during 2013–2018. Data on primary and natural forest were retrieved from Global Forest Watch (http://www.globalforestwatch.org). Right y-axis represents number of total number of equids tested for seroprevalence. Horizontal bars indicate means. Areas below dotted line had no forest loss. BSQV, Bussuquara virus; CPCV, Cacipacoré virus; ha, hectare (10,000 m2); PRNT90, 90% plaque-reduction neutralization test; ROCV, Rocio virus; SLEV, Saint Louis encephalitis virus; WNV, West Nile virus.

For antibody screening, we used an experimental WNV IgG ELISA based on a fusion loop envelope antigen containing mutations. We chose this ELISA to decrease the chances of cross-reactivity with antibodies elicited by other flaviviruses (2). Among 713 serum samples, 47 (6.6%, 95% CI 4.9%-8.7%) yielded positive ELISA results (Figure 2, panel A). Beyond WNV, horses in Latin America frequently are infected with Saint Louis encephalitis virus (SLEV), Cacipacoré virus (CPCV), Rocio virus (ROCV), and Bussuquara virus (BSQV) (12); and WNV, CPCV, and SLEV all belong to the Japanese encephalitis serocomplex (Appendix Figure 1, https://wwwnc.cdc.gov/EID/article/27/9/20-4706-App1.pdf). Serologic analyses for WNV-specific antibodies in horses could be compromised by crossreactive antibodies against other flaviviruses, eliciting potentially false-positive test results (11). Therefore, we confirmed ELISA-based WNV antibody detection by comparing the endpoint titers for the 90% plaque-

Table 1. West Nile	e virus seroprevalend	ce per	municipality, Brazil
			% Seroprevalence
Municipalities	Sampling year(s)	No.	(95% CI)*
Antonio Cardoso	2015, 2016	10	0
Barreiras	2014, 2017, 2018	17	0
Caatiba	2018	19	0
Conceição do	2013	29	0
Jacuípe			
Conde	2013	28	3.6 (0.1-18.4)
Cotegipe	2013	11	0
Cristopolis	2013	10	0
Esplanada	2013	57	3.5 (0.4-12.1)
Eunápolis	2013, 2014	21	0
Feira de santana	2013	29	0
Formosa do Rio	2013, 2017	37	0
Preto			
Gongogi	2018	23	4.3 (0.1-21.9)
Ibotirama	2013	6	0
Igaporã	2013	27	0
Itabela	2013	6	0
Itabuna	2013, 2017	41	2.4 (0.1-12.9)
Itaju do Colônia	2013, 2015	6	0
Itapetinga	2018	14	7.1 (0.2-33.9)
Jaborandi	2017	5	20.0 (0.5-71.6)
Juazeiro	2013, 2017	49	2.0 (0.5-14.0)
Lauro de Freitas	2017	14	0
Mata de São	2015, 2016, 2017	11	9.1 (0.2-41.3)
João			
Mucuri	2013	13	0
Palmas de	2013	18	0
Monte Alto			
Riachão das	2017, 2018	122	13.9 (8.3-21.4)
Neves			
Rio Real	2013	25	0
Serra Dourada	2017	6	16.7 (0.4-64.1)
Ubaitaba	2018	7	42.9 (9.9–81.6)
Others†	2013-2018	52	3.8 (0.5-13.2)
Total	2013-2018	713	4.5 (3.1-6.3)

^{*}Seroprevalence is based on 90% endpoint plaque-reduction neutralization tests.

reduction neutralization tests (PRNT₉₀), considered the standard for arbovirus serologic testing, for WNV, CPCV, SLEV, BSQV, and ROCV in all 47 ELISA-positive serum samples. Of the 47 samples, 20 (44.7%) neutralized WNV only in PRNT on; another 22 (46.8%) showed heterotypic reactions for WNV, CPCV, or SLEV (Figure 2, panel B). Averaged endpoint titers were significantly higher for WNV than for the other flaviviruses (p<0.0001) and exceeded those for CPCV, SLEV, BSQV, or ROCV by ≥4-fold in 12/22 heterotypic samples (Figure 2, panel C), a titer difference commonly considered decisive in flavivirus serology. Thus, 68.1% (32/47) of ELISA-positive samples had WNV-specific antibody responses (Figure 2, panel C); 4 samples were seronegative for all 5 flaviviruses by PRNT_{oo}, potentially because of differential sensitivity of ELISA and PRNT. No samples had higher SLEV-, BSQV-, or ROCV-specific PRNT on titers compared with WNV, but 2 ELISA-positive samples had ≥4-fold endpoint titers for CPCV compared with WNV and other flaviviruses (Appendix Table 1). These findings substantiated WNV and CPCV cocirculation among equids in northeastern Brazil, which is consistent with previous data on CPCV circulation in another region of Brazil (12), and high specificity of the ELISA-based screening algorithm.

PRNT₉₀ validated the overall WNV seroprevalence of 4.5% (32/713 samples; 95% CI 3.1%-6.3%), which we used for downstream analyses (Table 1). We noted seropositive animals in 11/28 municipalities distributed over ≈900 km², suggesting wide geographic spread of WNV (Table 1; Figure 1). We observed a concentration of positive samples in 2018 (Appendix Figure 2), and in 1 municipality (Figure 1), comprising 9 different seropositive farms with an average seroprevalence of 13.9% (95% CI 8.3%-21.4%). Antibody levels typically decline over time after flavivirus infection (13), which might bias positivity rates of serologic tests. However, WNV-specific PRNT_{oo} endpoint titers were significantly lower in 2018 than in the preceding years by Mann-Whitney test (p = 0.002), excluding a bias from hypothetically more recent WNV infection in the animals sampled in 2018 (Appendix Figure 3).

We performed generalized linear model analyses and principal component analysis to compare 12 environmental, ecologic, and demographic factors potentially affecting WNV seroprevalence (Table 2; Appendix Figure 4). Anthropogenic changes of pristine habitats can increase the abundance of zoonotic pathogens (14), potentially including arboviruses like WNV (15). Indeed, the model considering forest loss, but not the model considering tree cover alone, was

[†]Detailed information, including municipalities with >5 serum samples, is available in Appendix Table 2 (https://wwwnc.cdc.gov/EID/article/27/9/20-4706-App2.pdf).

Table 2. Mathematical modeling of ecologic factors potentially affecting West Nile virus seroprevalence, Brazil*

						Maximum OR difference		
Model	AIC	∆AIC	AW	p value†	OR (95% CI)	among study sites‡	ρ§	Comment#
Distance to bird route, km								
Coastal	248.02	0.00	0.56	0.001	1.002 (1.001–1.004)	4.527	0.09	+
Northeastern	251.41	3.39	0.10	0.009	1.003 (1.001–1.006)	6.813	0.08	+
Central	252.17	4.16	0.07	0.014	0.999 (0.997-1.000)	4.545	-0.08	_
Forest loss, y/n	250.38	2.37	0.17	0.005	5.106 (1.518-31.796)	5.106	0.09	+
Presence of natural or	253.39	5.38	0.04	0.029	3.186 (1.111–13.48)	3.186	0.08	+
primary forest, y/n								
Altitude, m	255.53	7.51	0.01	0.105	1.002 (1.000-1.004)	3.518	0.06	+
Mean temperature, °C	258.03	10.01	0.00	0.719	0.876 (0.427-0.803)	1.613	-0.04	_
Hottest quarter	255.57	7.55	0.01	0.108	0.617 (0.347-1.113)	5.155	-0.04	_
Human density, no./km ²	255.76	7.74	0.01	0.121	1.000 (1.000–1.001)	3.137	-0.01	+
Tree cover, %	256.87	8.86	0.01	0.257	0.981 (0.941-1.012)	2.618	-0.09	_
Horse density, no./km2	258.10	10.09	0.00	0.817	0.969 (0.741-1.275)	1.170	-0.03	_
Mean precipitation, mm	258.15	10.14	0.00	0.948	1.000 (0.999-1.001)	1.047	-0.01	_

*Models are sorted by AIC, an estimator of the model's quality; models with lower AIC values are superior to models with higher AIC values. Horse and human densities were based on 2018 data available from the Brazilian Institute of Geography and Statistics (https://www.ibge.org.br). Information on precipitation and mean temperature was obtained from WorldClim version 2 (https://www.worldclim.org). Information on tree cover was obtained from Copernicus Global Land Cover (https://loviewer.vito.be/download). Information on natural or primary forest loss was obtained from Global Forest Watch (https://www.globalforestwatch.org). AIC, Akaike information criterion; AW, Akaike weight; OR, odds ratio; \(\Delta AIC, \text{ the difference between a given and the best-supported model in AIC.} \)

significantly associated with higher WNV seroprevalence (odds ratio [OR] 5.106, 95% CI 1.318–31.796; p = 0.005) (Table 2). Model results were consistent with a higher proportion of WNV-seropositive sites in disturbed areas compared with pristine areas by χ^2 test (p = 0.009) (Figure 2, panel D). Higher WNV seroprevalence in disturbed areas was not biased by the number of animals living in those sites compared to sites from pristine areas. By Student t-test, neither the overall number of animals nor the number of animals per site differed significantly between disturbed (p = 0.9) and pristine areas (p = 0.2894) (Figure 2, panel D; Appendix Figure 5).

Because the geographic distribution of both the 2018 horse epizootic and the only known human case might be linked geographically to the northeastern and coastal avian migratory routes (Figure 1), we included distances to avian routes in model analyses of WNV seroprevalence. Proximity to the central avian migratory route was associated with higher WNV seroprevalence (Table 2; Appendix Figure 4). This finding was consistent with other seroprevalence studies, indicating the presence of WNV in horses in the central region in Brazil (7,8), but failed to connect the WNV detections in Brazil to geographically adjacent avian migratory routes. Our data were consistent with prior studies of WNV ecology, but the explicatory power of our models was low despite statistical significance (Table 2; Appendix Figure 4).

Our study was limited by absence of longitudinal samples from individual sampling sites, lack of information on animal trade and animal age, and relatively low numbers of seropositive animals from individual sites. Thus, we cannot exclude biases affecting the accuracy of our modeling approach. However, our large sample and the combination of thorough serologic analyses and mathematical modeling enabled robust estimates of WNV spread that can guide prospective studies.

Conclusions

Our findings of substantial WNV seroprevalence in equids in Brazil warrants WNV surveillance in cases of acute neurologic disease in humans and horses. In addition, blood products should be screened in areas of Latin America with high risk for WNV.

Acknowledgments

We thank Sandra Junglen and Anne Kopp for providing the Saint Louis encephalitis virus strain; Xavier de Lamballerie for providing the Cacipacoré virus, Bussuquara virus, and Rocio virus strains; and Patricia Tscheak for technical assistance. We thank the Friedrich-Loeffler-Institut, Bundesforschungsinstitut für Tiergesundheit, Insel Riems for providing HD11 (CCLV-RIE 1510) cells.

This work was supported by the European Union's Horizon 2020 research and innovation programme through the ZIKAlliance project (grant no. 734548).

[†]p values were determined by likelihood ratio tests of the different models.

[‡]Maximum OR difference among study sites indicates the highest OR difference possible for a given variable for better comparability between binary and nonbinary variables.

 $[\]S \rho$, the Spearman correlation coefficient, ranges between -1 for negative correlations and 1 for positive correlations. The closer ρ is to 1 or -1, the greater the correlation between the observed variables.

[#]Clarification that the observed variable is associated with an increase (+) or a decrease (-) of West Nile virus prevalence.

DISPATCHES

About the Author

Dr. Oliveira-Filho is a virologist at the Institute of Virology, Charité Universitätsmedizin Berlin. His research interests include the epidemiology and evolution of arthropod-borne viruses from animal reservoirs. Dr. Fischer is a virologist at the Institute of Virology, Charité Universitätsmedizin Berlin. His research interests include the epidemiology of arthropod-borne viruses in humans and animal reservoirs.

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Address for correspondence: Dr. Jan Felix Drexler, Helmut-Ruska-Haus, Institute of Virology, Campus Charité Mitte, Charitéplatz 1, 10117 Berlin, Germany; email: felix.drexler@charite.de

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Appendix

Appendix Methods

Samples

During routine surveillance activities conducted by local animal health agencies during 2013–2018, a total of 713 serum samples were collected from 618 horses and 95 mules. Animals were sampled in a large geographic area comprising 61 of the 417 municipalities in the state of Bahia.

Serologic Detection

After initial ELISA-based screening, 47 positive serum samples were tested by using plaque-reduction neutralization tests with a 90% endpoint (PRNT₉₀). Samples were tested for specific neutralizing antibodies against West Nile virus (WNV) by using strain NY-99, Cacipacoré virus (CPCV) by using strain UVE/CPCV/1977/BR/Be An 327600, Saint Louis encephalitis virus (SLEV) by using strain MSI-7, Bussuquara virus (BSQV) by using strain BeAn 4073, and Rocio virus (ROCV) by using strain UVE/ROCV/1975/BR/5P H34 675. PRNT₉₀ was conducted in monolayers of 1.6×10^5 VeroFM cells for WNV and BSQV, 1.6×10^5 VeroB4 cells for ROCV, 3.5×10^5 HD11 (CCLV-RIE 1510) cells derived from chicken macrophages for CPCV, or 1.2 × 10⁵ baby hamster kidney 21 (BHK-21) cells for SLEV, seeded in 12-well plates 1 day before the infection. We incubated 50 plaque forming units (PFUs) with serum dilutions for 1 h and added to the cell monolayer. After the 1-hour incubation, we added an overlayer containing Dulbecco's Modified Eagle Medium (DMEM; Gibco, https://www.thermofisher.com) with 2% fetal calf serum (FCS; Gibco), and 2.5% carboxymethyl cellulose (CMC; Carl Roth, https://www.carlroth.com). The medium was removed after 2 days for WNV, after 4 days for SLEV, after 5 days for CPCV and BSQV, or after 6 days for ROCV; cells were fixed with 6% paraformaldehyde and stained with crystal violet. The endpoint titration of positive serum samples was done by testing 1:40, 1:80, 1:160, 1:320, 1:640, 1:1,280, 1:2,560 and 1:5,120 serum dilutions; we considered titers reducing the number of PFU by ≥90% positive. Titers were calculated using the logistic regression function in Prism 6 (GraphPad software, https://www.graphpad.com).

Principal Component Analysis

We conducted a principal component analysis (PCA) by using the complete sample dataset to visualize the structure of the dataset and the relationship of different variables. For the PCA, we removed any unneeded variables from the dataset. We used R version 4.0.2 (R Foundation for Statistical Computing, https://www.r-project.org) and factoextra version 1.0.7 (BEAR Applications, University of Birmingham, UK, https://bear-apps.bham.ac.uk) to conduct PCA. We used cosign-squared (cos²) values to color arrows for better visualization.

Generalized Linear Model Analyses

We calculated generalized linear models (GLMs) GLMs to compare the effects of environmental, ecologic, and demographic factors on the seroprevalence of WNV. After initial testing, we selected 12 variables for further analyses on the basis of Akaike information criterion (AIC) values and their reasonability; for example, we selected only the precipitation variable having the best AIC support. We calculated GLMs, the statistical support of the GLMs, and odds ratios (95% CI) in R version 4.0.2 (R Foundation for Statistical Computing) by using the glm() function and we conducted a Spearman correlation test by using the cor.test() function. We analyzed AIC values by using the bbmle version 1.0.23.1 package in R (R Foundation for Statistical Computing).

Appendix Table 1. Comparative plaque-reduction neutralization test results for West Nile virus ELISA-positive serum samples from equids, Brazil*

			ELISA	ELISA	_ISA _ PRNT ₉₀ endpoint titers			PRNT ₉₀	Sampling		
Sample ID	Species	Municipality	OD	ratio	WNV	CPCV	SLEV	BSQV	ROCV	interpretation	year
ZS-A-1367	Horse	Barreiras	0.521	1.4	1:14	1:20	<10	<10	<10	Undetermined	2018
ZS-A-1246	Horse	Caatiba	0.547	1.4	1:16	1:31	<10	1:15	<10	Undetermined	2018
ZS-A-19	Horse	Conde	0.953	2.5	1:343	1:41	1:14	1:15	<10	WNV	2013
ZS-A-224	Horse	Esplanada	1.001	2.6	1:50	1:296	<10	<10	<10	CPCV	2013
ZS-A-237	Horse	Esplanada	0.652	1.7	1:166	1:40	<10	<10	1:11	WNV	2013
ZS-A-239	Horse	Esplanada	0.582	1.5	1:307	1:76	<10	<10	1:12	WNV	2013
ZS-A-1201	Mule	Gongogi	0.525	1.4	1:33	<10	<10	<10	<10	WNV	2018
ZS-A-789	Horse	Itabuna	0.682	1.8	1:26	<10	<10	<10	<10	WNV	2013
ZS-A-1219	Horse	Itapetinga	0.995	2.6	1:356	1:40	<10	<10	<10	WNV	2018
ZS-A-1221	Horse	Itapetinga	0.453	1.2	<10	<10	<10	<10	<10	Negative	2018
ZS-A-819	Horse	Jaborandi	0.685	1.8	1:81	1:38	<10	1:26	1:20	Undetermined	2017
ZS-A-820	Horse	Jaborandi	0.534	1.4	1:83	1:15	1:15	<10	<10	WNV	2017
ZS-A-1108	Mule	Juazeiro	0.531	1.4	1:85	<10	<10	<10	<10	WNV	2013
ZS-A-481	Horse	Lapão	0.687	1.8	1:209	<10	<10	<10	<10	WNV	2014
ZS-A-125	Horse	Mata de São	0.618	1.6	1:361	1:41	<10	<10	<10	WNV	2015
		João									
ZS-A-127	Horse	Pojuca	0.734	1.9	1:143	<10	<10	<10	<10	WNV	2016

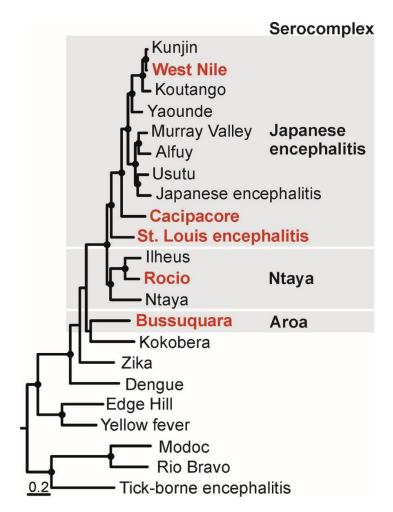
			ELISA	ELISA		DDNIT	₉₀ endpoi	nt titere		PRNT ₉₀	Sampling
Sample ID	Species	Municipality	OD	ratio	WNV	CPCV	SLEV	BSQV	ROCV	interpretation	Sampling year
ZS-A-1257	Horse	Riachão das Neves	1.458	3.8	1:57	<10	<10	<10	<10	WNV	2018
ZS-A-1258	Horse	Riachão das Neves	1.306	3.4	1:10	<10	<10	<10	<10	WNV	2018
ZS-A-1259	Horse	Riachão das	1.106	2.9	1:50	1:12	<10	<10	<10	WNV	2018
ZS-A-1263	Horse	Neves Riachão das	1.033	2.7	1:27	<10	<10	<10	<10	WNV	2018
ZS-A-1266	Horse	Neves Riachão das	0.983	2.6	1:78	<10	<10	<10	<10	WNV	2018
ZS-A-1267	Horse	Neves Riachão das	0.927	2.4	1:108	1:12	<10	1:26	<10	WNV	2018
ZS-A-1268	Horse	Neves Riachão das	0.876	2.3	1:38	<10	1:22	1:28	<10	Undetermined	2018
ZS-A-1272	Horse	Neves Riachão das Neves	0.84	2.2	1:11	<10	<10	<10	<10	WNV	2018
ZS-A-1276	Horse	Riachão das	0.833	2.2	<10	<10	<10	<10	<10	Negative	2018
ZS-A-1281	Horse	Neves Riachão das	0.81	2.1	1:44	<10	<10	<10	<10	WNV	2018
ZS-A-1283	Horse	Neves Riachão das	0.808	2.1	1:13	1:85	<10	<10	<10	CPCV	2018
ZS-A-1299	Horse	Neves Riachão das Neves	0.782	2.0	1:54	<10	<10	<10	<10	WNV	2018
ZS-A-1303	Horse	Riachão das Neves	0.76	2.0	1:19	<10	<10	<10	<10	WNV	2018
ZS-A-1304	Mule	Riachão das Neves	0.746	1.9	1:30	<10	1:19	<10	<10	Undetermined	2018
ZS-A-1307	Horse	Riachão das Neves	0.745	1.9	1:75	<10	<10	<10	<10	WNV	2018
ZS-A-1311	Horse	Riachão das Neves	0.731	1.9	1:52	1:24	<10	1:24	<10	Undetermined	2018
ZS-A-1316	Horse	Riachão das Neves	0.693	1.8	1:41	<10	<10	<10	<10	WNV	2018
ZS-A-1321	Horse	Riachão das Neves	0.688	1.8	1:19	<10	<10	<10	<10	WNV	2018
ZS-A-1322	Horse	Riachão das Neves	0.596	1.5	1:25	<10	<10	<10	<10	WNV	2018
ZS-A-1327	Horse	Riachão das Neves	0.555	1.4	1:432	<10	<10	<10	<10	WNV	2018
ZS-A-1328	Horse	Riachão das Neves	0.503	1.3	1:12	<10	1:16	<10	<10	Undetermined	2018
ZS-A-1329	Horse	Riachão das Neves	0.499	1.3	1:419	1:32	1:15	1:11	1:12	WNV	2018
ZS-A-1350	Horse	Riachão das Neves	0.49	1.3	<10	<10	<10	<10	<10	Negative	2018
ZS-A-1373	Mule	Riachão das Neves	0.489	1.3	1:74	<10	1:25	<10	<10	Undetermined	2018
ZS-A-1374	Mule	Riachão das Neves	0.484	1.3	<10	<10	<10	<10	<10	Negative	2018
ZS-A-1376	Mule	Riachão das Neves	0.443	1.1	1:13	<10	<10	<10	<10	WNV	2018
ZS-A-1378	Mule	Riachão das Neves	0.434	1.1	1:20	1:14	1:27	<10	<10	Undetermined	2018
ZS-A-842	Horse	Serra Dourada	0.66	1.7	1:329	<10	<10	<10	<10	WNV	2017
ZS-A-1186	Mule	Ubaitaba	1.414	3.7	1:310	<10	<10	1:18	<10	WNV	2018
ZS-A-1191	Mule	Ubaitaba	1.348	3.5	1:15	<10	<10	<10	<10	WNV	2018
ZS-A-1380	Horse	Ubaitaba	0.534	1.4	1:325	<10	1:14	<10	<10	WNV	2018

*BSQV, Bussuquara virus; CPCV, Cacipacoré virus; ID, identification; OD, optical density; PRNT₉₀, plaque-reduction neutralization test using a 90% endpoint; ROCV, Rocio virus; SLEV, Saint Louis encephalitis virus; WNV, West Nile virus.

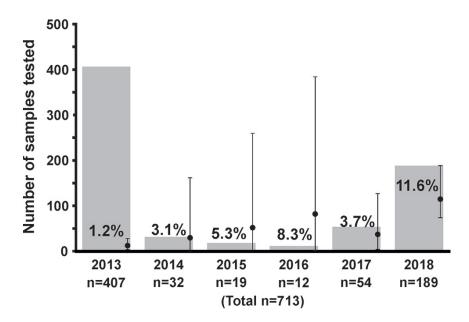
Appendix Table 2. West Nile virus seroprevalence among equids per municipality, Brazil % Seroprevalence

Municipalities Sampling year No. (95% CI)*				% Seroprevalence
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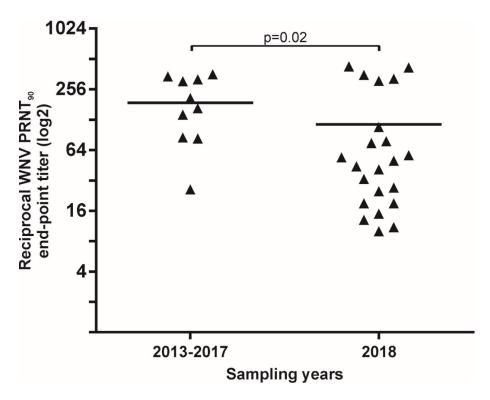
^{*}Based on plaque-reduction neutralization test using a 90% endpoint



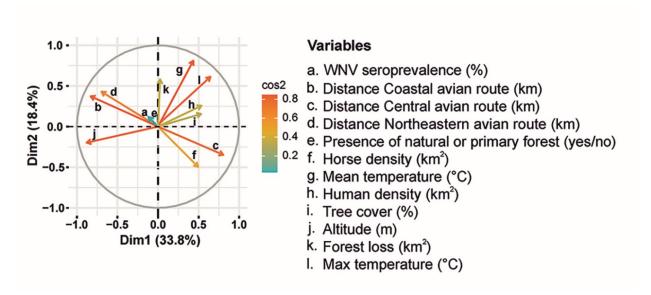
Appendix Figure 1. Maximum likelihood phylogenetic tree showing relevant flaviviruses used to assess West Nile virus seroprevalence among equids, Brazil. The tree was generated by using MEGA X (https://www.megasoftware.net) and a dataset comprising translated polyprotein genes and a Whelan and Goldman amino acid substitution model. Red text indicates flaviviruses from Brazil. Black dots indicate support values >0.70 from 500 bootstrap replicates. Scale bar indicates nucleotide substitutions per site.



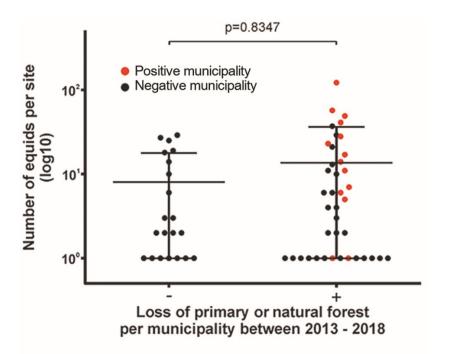
Appendix Figure 2. Seroprevalence of West Nile virus among equids over time, Brazil. Seroprevalence is based on plaque-reduction neutralization tests using a 90% endpoint (PRNT₉₀). Black dots indicate seroprevalence values of each year; vertical bars indicate 95% CI.



Appendix Figure 3. Comparison between 2018 and 2013–2017 West Nile virus–specific titers of plaque-reduction neutralization tests using a 90% endpoint (PRNT₉₀). Triangles represent PRNT₉₀–positive sera.



Appendix Figure 4. Principal component analysis in 2 dimensions (Dim) showing the relationship of variables used to model ecologic factors potentially affecting West Nile virus (WNV) seroprevalence, Brazil. Squared cosine (cos2) measures the interaction between 2 variables and is expressed in a gradient computing the effect of a principal component's contribution over a given observation.



Appendix Figure 5. Number of equids per municipality with (+) and without (–) primary or natural forest loss. Bars indicate mean and standard deviation. Statistical significance was determined by Student *t*-test.